

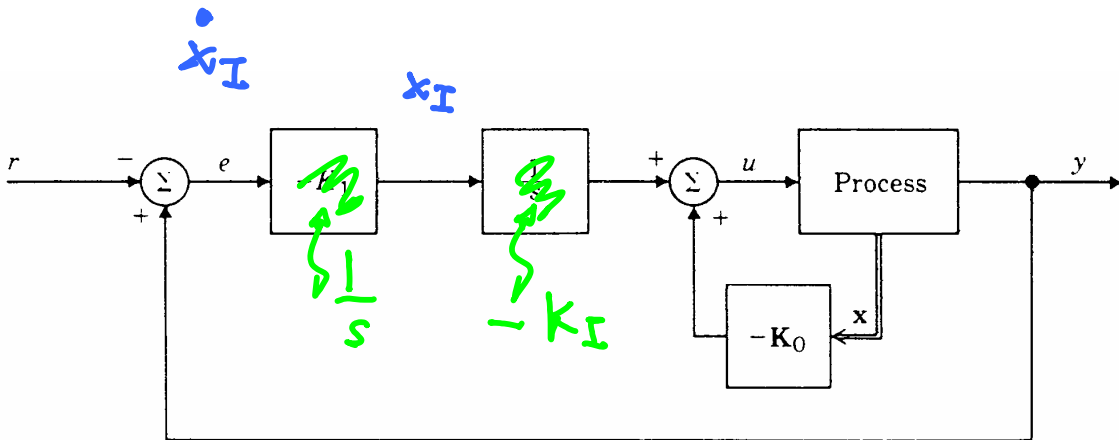
- choose your  $\alpha_c(s)$ .  
calculate control gain  $K$  ;
- if all states can be measured, skip this step.  
if states need to be estimated  
choose your  $\alpha_e(s)$ ;
- use the loop structure shown.  
use a package like Matrix<sub>x</sub> or MATLAB,  
find the transfer function from  $r^1 \mapsto y$ .
- for the t.f.  $r^1 \mapsto y$ , find the gain at  $s = j\omega = 0$ .  
Then use  $r^1 = K_s r$  where  $K_s$  is a scaling gain so  
that

$$K_s G_{r^1 \mapsto y}(s) \Big|_{s=j\omega=0} = 1$$

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- important because in every system, there is possibility of disturbance
- state feedback considered thus far will not reject a persistent disturbance
- inclusion of integral control also automatically takes care of steady state reference signal tracking.



Proper way to introduce integral action:

“state augmentation”

Plant

$$\dot{x} = Fx + Gu$$

$$y = Hx$$

Augment (i.e. add in) an extra state

$$\dot{x}_I = y - r$$

$$\dot{x}_I = y - r$$

Then, augmented system becomes

$$\begin{bmatrix} \dot{x} \\ \dot{x}_I \end{bmatrix} = \begin{bmatrix} F & 0 \\ H & 0 \end{bmatrix} \begin{bmatrix} x \\ x_I \end{bmatrix} + \begin{bmatrix} G \\ 0 \end{bmatrix} u$$

$$y = \begin{bmatrix} H & 0 \end{bmatrix} \begin{bmatrix} x \\ x_I \end{bmatrix}$$

or

$$\left. \begin{aligned} \dot{\bar{x}} &= \bar{F}\bar{x} + \bar{G}u \\ y &= \bar{H}\bar{x} \end{aligned} \right\}$$

Also consider

①

$$\dot{x} = \bar{F}x + Gu$$

$$+ G_I w$$

a constant disturbance

② Suppose we also need

$$[y(t) - r] \rightarrow 0$$

What then?

Now that this is in the standard state-space form, we can apply all the methods we talked out.

Example:

Plant  $\frac{Y(s)}{U(s)} = \frac{1}{s+3}$

- wish to have integral control with closed-loop poles at  $\omega_n = 5, \zeta = 0.5$

i.e.  $\alpha_c(s) = s^2 + 5s + 25$

- augmented system

$$\begin{bmatrix} \dot{x}_I \\ \dot{x} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & -3 \end{bmatrix} \begin{bmatrix} x_I \\ x \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u + \begin{bmatrix} -1 \\ 0 \end{bmatrix} r$$

(should not bother you by now whether

$$\bar{x} = [x_I \quad x]^T \text{ or } \bar{x} = [x \quad x_I]^T !!$$

You should be able to keep track of the states at this stage!!)

- calculate your control gains  $K$  from

$$\det[sI - \begin{bmatrix} 0 & 1 \\ 0 & -3 \end{bmatrix} + \begin{bmatrix} 0 \\ 1 \end{bmatrix} K] = s^3 + 5s + 25$$

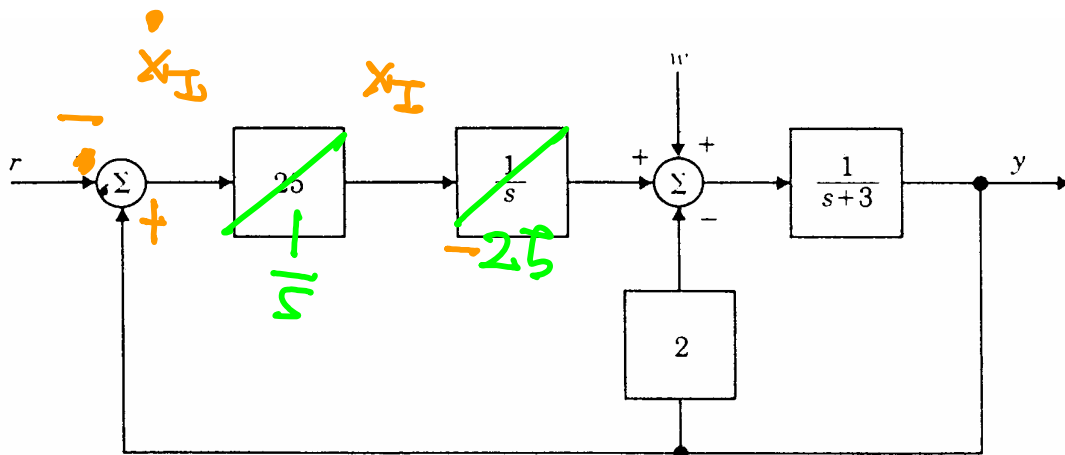
This gives

$$K = \begin{bmatrix} 25 & 2 \end{bmatrix}$$

and  $u = -K \begin{bmatrix} x_I \\ x \end{bmatrix}$  in this case.

Integral control example Block diagram:

- zero steady state error to step  $r$
- zero steady state error to constant (persistent) disturbance  $w$ .



## An Appendix on Solution of State Equations

[Not for exams]

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System  $\dot{x} = Fx + Gu$   
 $y = Hx$

Consider first  $u(t) \equiv 0$ , with  $x(0) = x_o$

Define a matrix exponential function by the power series

$$e^{Ft} = (I + Ft + \frac{F^2 t^2}{2!} + \dots)$$

$$= \sum_{K=0}^{\infty} \frac{F^K t^K}{K!} \quad \text{with} \quad F^0 = I$$

$$K! = 1$$

$$\dot{x} = Fx; \quad x(0) = x_o$$

Assume solution is given by

$$x(t) = e^{Ft} x_o$$

Check if this is correct


First note

$$\begin{aligned}
 \frac{d}{dt} e^{Ft} &= \frac{d}{dt} \left[ \sum_{K=0}^{\infty} \frac{F^K t^{K-1}}{K!} \right] \\
 &= \sum_{K=1}^{\infty} \frac{F^K t^{K-1}}{(K-1)!} \\
 &= F \sum_{j=0}^{\infty} \frac{F^j t^j}{j!} = F e^{Ft}
 \end{aligned}$$

Now let us see if solution correct.

- at  $t = 0$ ,

$$e^{Ft} x_o = e^{F \cdot 0} x_o = I x_o = x_o$$


  
 using def<sup>n</sup> of  $e^{Ft}$  with  $t = 0$

$\therefore$  fits at  $t = 0$

- Check if  $\dot{x} = Fx$

$$\begin{aligned}
 \text{L.H.S.} = \dot{x} &= \frac{d}{dt} (x(t)) = \frac{d}{dt} (e^{Ft} x_o) \\
 &= F e^{Ft} x_o \quad \text{from above} \\
 &= Fx = \text{R.H.S.}
 \end{aligned}$$

$\therefore$  fits state-space d.e. as well.

$\therefore$  solution to

$$\dot{x} = Fx; \quad x(0) = x_o$$

is  $x(t) = e^{Ft} x_o$

$\Delta\Delta$

What about case with non-zero  $u(t)$ ?

$$\dot{x} = Fx + Gu; \quad x(0) = x_o$$

Solution is in fact

$$x(t) = e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau$$

Compare this  
with your  
"A"-level, JC,  
or high school  
mathematics!

Solution confirmed in same way:

- at  $t = 0$

$$t = 0, \quad e^{Ft} x_o = x_o \quad (\text{from previous})$$

$$t = 0, \quad \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau = \int_0^0 [\quad] d\tau = 0$$

$\therefore$  fits at  $t = 0$

$$\begin{aligned} \dot{x} &= Fx + Gu \\ x(0) &= x_o \end{aligned}$$

$$x(t) = e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau$$



- Check if  $\dot{x} = Fx + Gu$

$$\begin{aligned} \text{L.H.S.} \quad &= \dot{x} = \frac{d}{dt} x(t) \\ &= \frac{d}{dt} \left[ e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau \right] \end{aligned}$$

remember that  $\frac{d}{dt} \int_0^t \alpha(t, \tau) d\tau$

$$= \alpha(t, \tau) \Big|_{\tau=t} + \int_0^t \frac{d}{dt} \alpha(t, \tau) d\tau$$

$$\begin{aligned} \therefore \frac{d}{dt} \left[ e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau \right] \\ &= Fe^{Ft} x_o + e^{F(t-\tau)} Gu(t) \Big|_{\tau=t} + \int_0^t Fe^{F(t-\tau)} Gu(\tau) d\tau \\ &= F \left[ e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau \right] + Gu(t) \\ &= Fx(t) + Gu(t) = R.H.S. \end{aligned}$$

$\therefore$  state-space equation is satisfied

Thus for

$$\dot{x} = Fx + Gu; \quad x(0) = x_o$$

$$y = Hx$$

Solution is

$$x(t) = e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau$$

and

$$y(t) = Hx(t)$$

$$= \underbrace{He^{Ft} x_o}_{\text{zero input response}} + \underbrace{\int_0^t He^{F(t-\tau)} Gu(\tau) d\tau}_{\text{zero state response}}$$

zero input  
response

zero state  
response

Note that when we considered impulse response,  
We said, let

$$\frac{Y(s)}{U(s)} = G(s) \quad \text{convolution}$$

then  $y(t) = \int_0^t g(t-\tau)u(\tau) d\tau = g(t) * u(t)$

where  $g(t)$  is impulse response.

From above, impulse response  $= g(t) = He^{Ft} G$

$\Delta\Delta$

## Ways to calculate $e^{Ft}$ (Franklin & Powell pp. 410)

- important previously
- not so important nowadays with advent of inexpensive computing facilities, and “canned” control system software

### 1. Matrix Exponential Series

use the def<sup>n</sup> directly

$$e^{Ft} = I + Ft + \frac{F^2 t^2}{2!} + \frac{F^3 t^3}{3!} + \dots$$

- can only be used for F with all negative e-values
- truncate when high-order terms sufficiently small

### 2. Inverse Laplace Transform

$$e^{Ft} = L^{-1} \{ (sI - F)^{-1} \}$$

- invert the matrix  $(sI - F)$ ;
- find the inverse Laplace transform for each element  $(sI - F)^{-1}_{ij}$

### 3. Diagonalization of System Matrix F

- can only be used if F is diagonalizable

find transformation T so that

$$T\Lambda = FT \text{ or } \Lambda = T^{-1}FT$$

↗

$$\begin{bmatrix} t_1 & t_2 & \dots & t_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & & 0 \\ & \lambda_2 & & \\ & & \ddots & \\ 0 & & & \lambda_n \end{bmatrix} = F \begin{bmatrix} t_1 & t_2 & \dots & t_n \end{bmatrix}$$

Optional further investigation:  
Generalizes to the "Jordan" matrix for repeated eigenvalues

thus columns of T are in fact eigenvectors of F, and  $\lambda_i$  are corresponding eigenvalues

(use a CAD package, if possible, to find T)

then

$$e^{Ft} = T \begin{bmatrix} e^{\lambda_1 t} & & & 0 \\ & e^{\lambda_2 t} & & \\ & & \ddots & \\ 0 & & & e^{\lambda_n t} \end{bmatrix} T^{-1}$$

Not for exams!

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## Digital Control: State-Space Methods

Recall that there are two main design philosophies in digital control

- (a) Discrete-time implementation of continuous-time controllers
- (b) Direct design of digital controllers

### (a) Discrete-Time Implementations of Continuous-Time Controllers

- Straightforward  
use rules like
  - Forward Rectangular Rule
  - Backward Rectangular Rule
  - Bilinear Transformation
  - Matched Pole-Zero Approximation
- sampling interval,  $h$ , must be chosen carefully  
typically  $\frac{1}{10}$  of smallest time constant in system

Each of these methods have advantages & disadvantages.

If  $h$  chosen as suggested (i.e.  $\frac{1}{10}$  of smallest time constant in system), then all the above methods can be used in the discrete-time implementation of the continuous-time controller.

Example using Forward Rectangular Rule (also called Delta operator).

- Some notation first.

use the notation  $p = \frac{d}{dt}$ , the differential operator, where:  
 i.e.  $px(t) = \frac{d}{dt}x(t)$

and introduce the forward-shift operator,  $q$ , where:  
 $qx(kh) = x(k+1h)$ ,  $k = 0, 1, 2, \dots$

note that  $p$  and  $q$  are time-domain operators  
 corresponding frequency domain variable are

$p$	$\rightarrow$	$s$
$q$	$\rightarrow$	$z$
time domain		frequency (or complex variable) domain

forward rectangular rule:

substitute  $\frac{z-1}{h}$  for  $s$

equivalently, in time domain

substitute  $\frac{q-1}{h}x(kh)$  for  $\frac{d}{dt}x(t)$   
at  $t = kh$

i.e.  
use  
 $\frac{x(\overline{k+1}h) - x(kh)}{h}$   
for  $\frac{d}{dt}x(t)$

Discrete-time implementation of controller in  
Franklin & Powell pp361 example

Plant is  $G(s) = \frac{1}{s^2}$

i.e.  $\dot{x} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u$

$$y = \begin{bmatrix} 1 & 0 \end{bmatrix} x$$

control roots to be placed at

$$s = (-1 \pm j1) / \sqrt{2} \quad (\omega_n = 1, \zeta = 0.7)$$

this gives  $K = \begin{bmatrix} 1 & \sqrt{2} \end{bmatrix}$

Estimator error roots are at

$$s = -2.5 \pm j4.3 \quad (\omega_n = 5, \zeta = 0.5)$$

this gives 
$$L = \begin{bmatrix} 5 \\ 25 \end{bmatrix}$$

Thus the continuous-time controller is

$$\begin{aligned} \dot{\hat{x}} &= (F - GK - LH)\hat{x} + Ly \\ &= \left\{ \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} - \begin{bmatrix} 0 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & \sqrt{2} \end{bmatrix} - \begin{bmatrix} 5 \\ 25 \end{bmatrix} \begin{bmatrix} 1 & 0 \end{bmatrix} \right\} \hat{x} \\ &\quad + \begin{bmatrix} 5 \\ 25 \end{bmatrix} y \end{aligned}$$

$$\dot{\hat{x}} = \begin{bmatrix} -5 & 1 \\ -26 & -\sqrt{2} \end{bmatrix} \hat{x} + \begin{bmatrix} 5 \\ 25 \end{bmatrix} y$$

$$u = -\begin{bmatrix} 1 & \sqrt{2} \end{bmatrix} \hat{x}$$

The time constants correspond to roots  
at  $s = 0$  (plant)

$$\tau \approx \frac{1}{\omega_n}$$

$$s = (-1 \pm j1) / \sqrt{2} \quad (\omega_n = 1, \zeta = 0.7) \text{ (control)} \quad \tau_c \approx 1$$

$$s = -2.5 \pm j4.3 \quad (\omega_n = 5, \zeta = 0.5) \text{ (estimation)} \quad \tau_e \approx \frac{1}{5}$$



$\therefore$  choose  $h$  to be  $\frac{1}{10}$  of smallest time constant

i.e.  $h = \frac{1}{10} \left( \frac{1}{5} \right) = 0.02$  seconds

$$\dot{\hat{x}} = \begin{bmatrix} -5 & 1 \\ -26 & -\sqrt{2} \end{bmatrix} \hat{x} + \begin{bmatrix} 5 \\ 25 \end{bmatrix} y$$

$$u = - \begin{bmatrix} 1 & \sqrt{2} \end{bmatrix} \hat{x}$$

discrete-time implementation:

$$\dot{\hat{x}}(t) \cong \frac{q-1}{h} \hat{x}(kh) = \frac{\hat{x}(\overline{k+1h}) - \hat{x}(kh)}{h}$$

$$\therefore \frac{q-1}{h} \hat{x}(kh) = \begin{bmatrix} -5 & 1 \\ -26 & -\sqrt{2} \end{bmatrix} \hat{x}(kh) + \begin{bmatrix} 5 \\ 25 \end{bmatrix} y(kh)$$

i.e.

$$h = 0.02$$

$$\hat{x}(\overline{k+1\{0.02\}})$$

$$= \begin{bmatrix} 1 - 5 \times 0.02 & 0.02 \\ -26 \times 0.02 & 1 - 0.02 \times \sqrt{2} \end{bmatrix} \hat{x}(0.02k)$$

$$+ \begin{bmatrix} 5 \times 0.02 \\ 25 \times 0.02 \end{bmatrix} y(0.02k) \quad \#$$

$$k = 0, 1, 2, 3, \dots$$

$$u(0.02k) = - \begin{bmatrix} 1 & \sqrt{2} \end{bmatrix} \hat{x}(0.02k) \quad \#$$

- thus design is carried out in continuous-time
- controller is implemented digitally in discrete-time, with  $\hat{x}$  and  $u$  updated every  $h = 0.02$  seconds.

$\Delta\Delta$

## (b) Direct Design of Digital (State-Space) Controllers

This requires that we first obtain a discrete-time state-space representation of the plant.  
(Franklin & Powell pp 562~567)

### Obtaining Discrete-Time State Representation

Plant  $\dot{x} = Fx + Gu \quad ; \quad x(0) = x_o$   
 $y = Hx$

Use sampling interval  $h$

Recall that solution to state-equation is

$$x(t) = e^{Ft} x_o + \int_0^t e^{F(t-\tau)} Gu(\tau) d\tau$$

- assume a zero-order hold

then  $u(\tau) = u(kh) \quad kh \leq \tau < (k+1)h \quad ; \quad k = 0, 1, 2, 3, \dots$

or, more generally =

$$x(t) = e^{F(t-t_0)} x(t_0) + \int_{t_0}^t e^{F(t-\tau)} G u(\tau) d\tau$$

then, using the solution to the state-equation,  
 $x(kh)$  may be related to  $x(kh+h)$  as

$$\tau = \overline{k+1} h$$

$$t_0 = kh$$

$$x(kh+h) = e^{Fh} \underline{x(kh)} + \int_{kh}^{kh+h} e^{F(kh+h-\tau)} d\tau \underline{Gu(kh)}$$

$$y(kh) = \underline{Hx(kh)}$$

and the discrete-time state-equation takes the form:

$$x(k+1) = \Phi x(k) + \Gamma u(k)$$

$$y(k) = Hx(k) \quad k = 0, 1, 2, \dots$$

cf  $\dot{x} = Fx + Gu$   
 $y = Hx$

where the sampling interval  $h$  is assumed

i.e.  $x(k)$  means  $x(kh)$

[for simplicity of notation!]

and  $\Phi = e^{Fh}$

$$\Gamma = \int_{kh}^{kh+h} e^{F(kh+h-\tau)} d\tau G$$

$$= \int_0^h e^{F\eta} d\eta G$$

for the subst<sup>n</sup> of  
variables  $\eta = kh + h - \tau$

Check this!

Thus, the discrete-time state-variable theory proceeds in exactly the same way as the continuous-time case, *using the discrete-time state-variable description:*

$$x(k+1) = \Phi x(k) + \Gamma u(k)$$

$$y(k) = Hx(k)$$

*compare with*

$$\dot{x} = Fx + Gu$$

$$y = Hx$$

Poles of discrete-time s.s. system: "a pole is a value of discrete-time frequency such that the system equation has a non-trivial solution when the forcing input is zero."

i.e. without a forcing input, the system responds as

$$x(k) = \alpha^k x_o \quad \text{for i.c.} \quad x(0) = x_o$$

then  $x(k+1) = \Phi x(k) + \Gamma u(k)$

i.e.  $\alpha^{k+1} x_o = \Phi \alpha^k x_o$

$$\alpha^k \{\alpha I - \Phi\} x_o = 0$$

$u \equiv 0$

*where  $\alpha$  is a pole of the system in the z-plane*

For a non-trivial solution ( $x_o \neq 0$ )

must have  $\det[\alpha I - \Phi] = 0$

i.e. poles,  $z$ , are the roots of  
 $\det[zI - \Phi] = 0$

(cf. c.t. s.s. poles are roots of  $\det[sI - F] = 0$ )

## State Feedback

consider control

$$u(k) = -Kx(k)$$

then overall system is

$$\begin{aligned} x(k+1) &= \Phi x(k) + \Gamma u(k) \\ &= [\Phi - \Gamma K] x(k) \end{aligned}$$

Compare:

$$\begin{aligned} \dot{x} &= f x + G u \\ &= (F - G K) x \end{aligned}$$

for  $u = -Kx$

$\therefore$  the poles of the system are arbitrarily assignable  
 if  $\zeta(\Phi, \Gamma)$  has full rank.

- Exact correspondence carries over from continuous-time

## State-Estimation

Consider full-order estimator

$$\hat{x}(k+1) = \Phi \hat{x}(k) + \Gamma u(k) + \underbrace{L}_{\text{orange}} [y(k) - H \hat{x}(k)]$$

State-error dynamics is,  $\tilde{x} = \hat{x} - x$ ,

$$\tilde{x}(k+1) = [\Phi - \underbrace{LH}_{\text{orange}}] \tilde{x}(k)$$

and arbitrary estimator dynamics  $\alpha_e(z)$  can be specified if  $O(H, \Phi)$  is full-rank.

Thus, once the sampled state-space equation in discrete-time directly is obtained, the controller can be designed directly in discrete-time using state-feedback & state-estimation.

Compare =

$$\dot{\hat{x}} = F \hat{x} + G u + \underbrace{L}_{\text{orange}} (y - H \hat{x})$$

and

$$\dot{\tilde{x}} = (F - \underbrace{LH}_{\text{orange}}) \tilde{x}$$